

CS229r: IT GCS
LECTURE 16

3/28/2019

TODAY

- ① Conclude: Compressing Interactions
- ② Direct Product // Amortized Communication.

————— γ —————

Review from last time

- ① Theorem: (to be proved): π can be compressed to length

$O(\sqrt{I \cdot k} \log k)$, if $I = IC(\pi)$, $k = CC(\pi)$.

- ② Ingredient: Correlated Sampling protocol.

Alice $\leftarrow P$; Bob $\leftarrow Q$; P, Q dist on Σ

Both get R ; should output $w_A \sim P$, $w_B \sim Q$

with $\Pr[w_A \neq w_B] \leq 2\delta(P, Q)$.

Protocol: $R = (a_1, b_1), \dots (a_i, b_i) \dots$

~~$a_i \in \Sigma, b_i$~~ , $a_i \sim \text{Unit}(\Sigma)$, $b_i \sim \text{Unit}([0, 1])$

• Alice outputs a_i where i is smallest index s.t. ~~$b_i \leq P(a_i)$~~

Protocol for Π & Priors

at level i
 for every node U_i in Protocol tree of Π can
 associate distributions

$$P_U \approx \Pi_i | (\Pi_{\leq i}, X, Y)$$

$$P_U^A \approx \Pi_i | (\Pi_{\leq i}, X)$$

$$P_U^B \approx \Pi_i | (\Pi_{\leq i}, Y)$$

~~Choices~~ ~~not yet~~ \longrightarrow

Rough Claim: if $P_U^A \approx \text{Bern}(\frac{1}{2} - \delta)$; $P_U^B \approx \text{Bern}(\frac{1}{2})$

then $I(C) = R\delta^2$; $\Pr[\text{Alice + Bob disagree on leaf}]$

$$= O(k\delta)$$

$$= O(\sqrt{R} \cdot \sqrt{I})$$

\longrightarrow

Next full Protocol + analysis

- ① Alice & Bob chose (correlatedly), nodes out of U for every U in tree.
- ② find first point U_1 on tree where they disagree
- ③ Repeat from U_1 ... until agree on leaf.

Analysis

① finding first point of disagreement = $O(\log k)$ bits of comm.
 \Rightarrow cost per iteration = $O(\log k)$

② # iterations = ?

• let Z_i = indicator variable indicating iteration starts at level i .

$$\cdot \Pr[Z_i=1] \leq \mathbb{E}_{\mathcal{U}}[\delta(P_U^A, P_U^B)]$$

where \mathcal{U} is random node at level i .

• Cost of i^{th} buyer = v_i

$$\approx \mathbb{E}_{\mathcal{U}}[\delta(P_U^A \| P_U^B)]$$

• Rest of analysis: convexity etc.

$$\# \text{ iterations} = \sum_i \mathbb{E}_{\mathcal{U}}[\delta(P_U^A, P_U^B)]$$

$$= \sum_i \mathbb{E}_{\mathcal{U}}\left[\sqrt{\delta(P_U^A \| P_U^B)}\right]$$

$$\leq \sqrt{R} \cdot \sum_i \mathbb{E}_{\mathcal{U}} \delta(P_U^A \| P_U^B)$$

$$\approx \sqrt{R \cdot I}$$



Direct Product Problems

- Given comm. complexity problem $f: X \times Y \rightarrow \mathbb{R}$
 Then n -fold direct product of f is $f^{\otimes n}: X^n \times Y^n \rightarrow \mathbb{R}^n$
 with $f^{\otimes n}(x_1, \dots, x_n, y_1, \dots, y_n) = (f(x_1, y_1), \dots, f(x_n, y_n))$.
- for distributional comm. problem, (f, μ) , the n -fold product
 in $(f^{\otimes n}, \mu^{\otimes n})$ $\left[\mu^{\otimes n}(x_1, \dots, x_n, y_1, \dots, y_n) = \prod_{i=1}^n \mu(x_i, y_i) \right]$.
- Trivial: $CC(f^{\otimes n}) \leq n \cdot CC(f)$.
- Question: Can $CC(f^{\otimes n})$ be smaller than $\Delta(n, CC(f))$?
- Note: protocol for $f^{\otimes n}$ doesn't have to work coordinatewise
 $\overbrace{\quad\quad\quad}^X \overbrace{\quad\quad\quad}^Y$
- Thm: [BBCR]: $\forall f \quad CC(f^{\otimes n}) \geq \sum (\sqrt{n} \cdot CC(f))$.
- Thm: [BR]: $\forall f \quad CC(f^{\otimes n}) = n \cdot IC(f)$ \star
- Thm: [GKR]: $\exists f \text{ s.t. } IC(f) = \log(CC(f))$
- Corollary: $\exists f \text{ s.t. } CC(f^{\otimes n}) \leq n \cdot \log(CC(f))$

\star with some careful definitions.

★ Careful Technical Definition

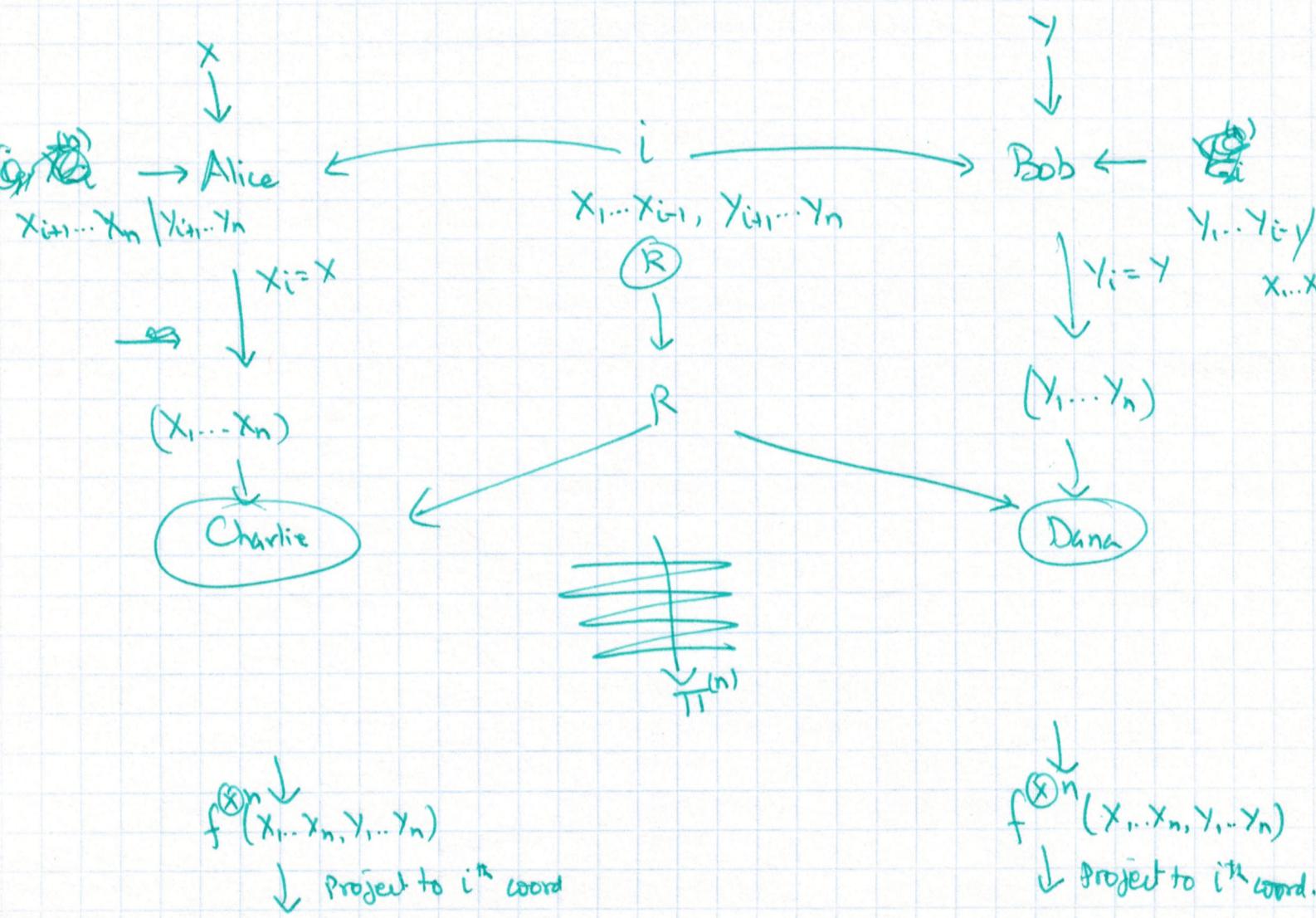
- How to deal with errors in randomized CC or $\{e\}$ direct product?
- we expect all coordinates to be right
- if $CC_{M, \epsilon}(f) \leq R$ then $CC_{M^n, \underbrace{1 - (1 - \epsilon)^n}_{\approx}}(f^{\otimes n}) \leq R \cdot n$
Not nice!
- Our definition: $\bar{\Pi}^{(n)}$ solves $f^{\otimes n}$ with error ϵ ,
if $\forall i \boxed{\bar{\Pi}^{(n)}(\bar{x}, \bar{y})_i = f^{\otimes n}(\bar{x}, \bar{y})_i \text{ w.p. } \geq 1 - \epsilon}$
- Nice: since now ~~now~~ we have natural ϵ -error protocol
gives n -fold ϵ -error protocol
- Not nice: since error of direct sum \neq error of protocol.
- But OK since we are mostly proving lower bounds
- $CC_{\epsilon}^n(f) \triangleq \min_{\substack{\bar{\Pi} \text{ that} \\ \text{solves } f^{\otimes n} \\ \text{with } \epsilon\text{-error}}} \{CC(\bar{\Pi})\}$; $IC_{\epsilon}^n(f)$ similar.

$$\text{Lemma 1: } \text{IC}_G^n(f) = n \cdot \text{IC}_G(f)$$

Proof: \leq obvious. Will prove \geq . By simulation.

- Suppose $\Pi^{(n)}$ solves $f^{\otimes n}$ with error ϵ , & $\text{IC} = I$
- Will show how to use $\Pi^{(n)}$ to construct Π that solves f with error ϵ & $\text{IC} = \frac{I}{n}$.

Embedding



7

Claims: ① Output correct w.p. $1-\epsilon$ ✓ (By Construction)

② $I_{\text{C}}(\pi) = \frac{I_{\text{C}}(\pi^n)}{n}$



$$\frac{1}{n} \sum_{i=1}^n I(x_i; \pi^{(n)} | x_1 \dots x_{i-1}, y_{i+1} \dots y_n, R; Y_i) \\ + I(Y_i; \pi^{(n)} | x_1 \dots x_{i-1}, y_{i+1} \dots y_n, R; X_i)$$



$$\frac{1}{n} (I(x_1 \dots x_n; \pi^{(n)} | R; Y_1 \dots Y_n) + I(Y_1 \dots Y_n; \pi^{(n)} | R; X_1 \dots X_n))$$

Are these equal? Yes! Chain Rule!

~~$x_i (x_1 \dots x_{i-1}, y_{i+1} \dots y_n, R, Y_i) - \pi^{(n)}$~~

+ the fact that $x_i = (x_1 \dots x_{i-1}, y_{i+1} \dots y_n, R, Y_i) - (\pi^{(n)}, Y_1 \dots Y_n)$

$$\Rightarrow \cancel{\sum_{i=1}^n I(x_i; \pi^{(n)} | x_1 \dots x_{i-1}, y_{i+1} \dots y_n, R, Y_i)}$$

$$= \sum_{i=1}^n I(x_i; \pi^{(n)} | x_1 \dots x_{i-1}, Y_1 \dots Y_n, R)$$

$$= I(x_1 \dots x_n; \pi^{(n)} | Y_1 \dots Y_n; R)$$



Theorem: $CC(f^{\otimes n}) \geq \frac{J_n \cdot CC(f)}{\log CC(f^{\otimes n})}$

Proof: Take protocol with best comm. complexity C for f .

- Yield protocols with $CC \leq IC + \frac{C}{n}$ for f .
- Compose to get protocol with $CC \leq \log \frac{C}{\sqrt{n}}$ for f .

$$\Rightarrow CC(f) \leq \frac{CC(f^{\otimes n})}{\sqrt{n}} \cdot \log C$$

[Concludes BBR]

Towards Braverman-Rao: $CC(f^{\otimes n}) = n \cdot IC(f)$

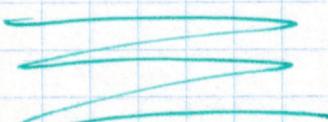
• Since $IC_G^n(f) = n \cdot IC_G(f)$ & $IC_E^n(f) \leq CC(f^{\otimes n})$

we get $CC(f^{\otimes n}) \geq n \cdot IC(f)$.

- To need to show \leq . Another "compression" result.
- Central Ingredient: Interactive Correlated Sampling

Alice $\leftarrow P$

$Q \rightarrow$ Bob



\downarrow
 $X \sim P$

\downarrow
 $Y = x \text{ w.h.p. } (\geq 1 - \epsilon)$

Comm: $D(P||Q) + 5\sqrt{D(P||Q)} + O(\log_{\frac{1}{\epsilon}} \frac{1}{\epsilon})$

⑨

- Contrast with last lecture: ① No interaction in last lecture.
- ② Here: Interaction + low error (~~high~~ ^{allows high} $D(P \parallel Q)$).

'Rough idea: Proof: NEXT LECTURE